

Biologically Inspired Computing

EECS 6180

Homework 6

Neural Evolution

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Introduction

A 4 layer NN and 5 layer NN were evolved to compare to the *Iris_bp* algorithm. The programs for all algorithms are provided in the appendix. Figure 1 is a sample NN with only 3 hidden neurons. Figure 2 and 3 represent a sample chromosome corresponding to the 4 and 5 hidden neuron evolutionary NN's designed for this project.

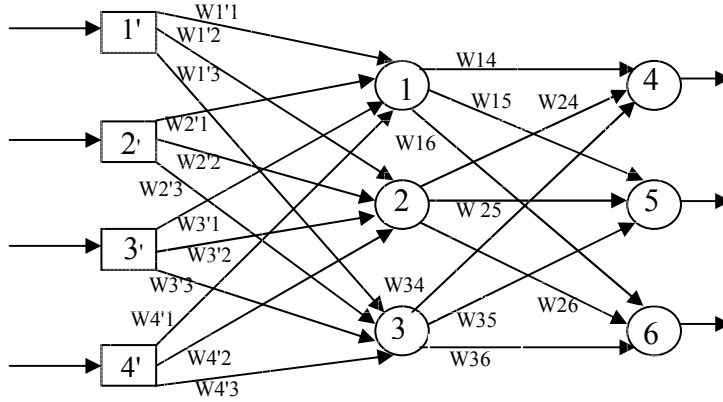


Fig. 1: Sample Neural network Architecture for Iris Data Set

W1'1	W1'2	W1'3	W1'4	W2'1	W2'2	W2'3	W2'4	W3'1	W3'2	W3'3	W3'4	W4'1	W4'2	W4'3	W4'4	W15	W16	W17	W25	W26	W27	W35	W36	W37	W45	W46	W47	T1	T2	T3	T4	T5	T6	T7
------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	----	----	----	----	----	----	----

Fig. 2: Chromosome 4 hidden neuron NN

W1'1	W1'2	W1'3	W1'4	W1'5	W2'1	W2'2	W2'3	W2'4	W2'5	W3'1	W3'2	W3'3	W3'4	W3'5	W4'1	W4'2	W4'3	W4'4	W4'5	W16	W17	W18	W26	W27	W28	W36	W37	W38	W46	W47	W48	W56	W57	W58	T1	T2	T3	T4	T5	T6	T7	T8
------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	------	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	-----	----	----	----	----	----	----	----	----

Fig. 3: Chromosome 5 hidden neuron NN

Technique

1. Training

In each case a set of 100 weight matrices were generated for Evolutionary training. This was done with:

```
W=rand(100, 35); for 4 hidden Neurons
W=rand(100, 43); for 5 hidden Neurons
```

The training and testing data were randomly selected from *Iris_data*. The training algorithm was a for loop which used a feed forward technique with a sigmoid function. The error for each input and subsequent set of weights (individual chromosome) was used to select the top 40 weight chromosomes.

2. Crossover

The top 40 weight chromosomes were crossed such that only weights corresponding to hidden1, hidden2..., hidden5, out1, out2, and out3 crossed with one another and genes from hidden1 did not cross with any other neuron's genes.

3. Mutation

The top 60 weight chromosomes were used for mutation and each individual gene had a probability of mutation = 0.1%

After each training generation if the mse of the training output was less than 0.07 then training would stop.

4. Testing

After either the training had run out of training samples or the training criteria was met then the testing was done in the same fashion only using the best weight chromosome determined from training.

Results

Method	Training time (s)	Epochs generations	Testing Error		
			%Error Iris-setosa	%Error Iris-versicolor	%Error Iris-verginica
Iris_bp	6.1347	1000	0%	10%	0%
Evo NN 4hidden	6.65	100	40%	100%	100%
Evo NN 5hidden	6.45	99	6.25%	0%	100%

Conclusion

The results show that there is more time needed to develop a good working Evolutionary Neural Network, but it is possible to obtain a better performance with respect to time.

```
function [iris_data] = Iris_data()

% =====
% Data set for Iris plant classification
% =====

% =====
% Reference: Negnevitsky, M., "Artificial Intelligence: A Guide to Intelligent
% Systems", 2nd edn. Addison Wesley, Harlow, England, 2005.
% Sec. 9.4 Will a neural network work for my problem?
% =====

% Iris plant data set

iris_data = [5.1 3.5 1.4 0.2 4.9 3.0 1.4 0.2 % Iris-setosa
             4.7 3.2 1.3 0.2 4.6 3.1 1.5 0.2 % Iris-setosa
             5.0 3.6 1.4 0.2 5.4 3.9 1.7 0.4 % Iris-setosa
             4.6 3.4 1.4 0.3 5.0 3.4 1.5 0.2 % Iris-setosa
             4.4 2.9 1.4 0.2 4.9 3.1 1.5 0.1 % Iris-setosa
             5.4 3.7 1.5 0.2 4.8 3.4 1.6 0.2 % Iris-setosa
             4.8 3.0 1.4 0.1 4.3 3.0 1.1 0.1 % Iris-setosa
             5.8 4.0 1.2 0.2 5.7 4.4 1.5 0.4 % Iris-setosa
             5.4 3.9 1.3 0.4 5.1 3.5 1.4 0.3 % Iris-setosa
             5.7 3.8 1.7 0.3 5.1 3.8 1.5 0.3 % Iris-setosa
             5.4 3.4 1.7 0.2 5.1 3.7 1.5 0.4 % Iris-setosa
             4.6 3.6 1.0 0.2 5.1 3.3 1.7 0.5 % Iris-setosa
             4.8 3.4 1.9 0.2 5.0 3.0 1.6 0.2 % Iris-setosa
             5.0 3.4 1.6 0.4 5.2 3.5 1.5 0.2 % Iris-setosa
             5.2 3.4 1.4 0.2 4.7 3.2 1.6 0.2 % Iris-setosa
             4.8 3.1 1.6 0.2 5.4 3.4 1.5 0.4 % Iris-setosa
             5.2 4.1 1.5 0.1 5.5 4.2 1.4 0.2 % Iris-setosa
             4.9 3.1 1.5 0.1 5.0 3.2 1.2 0.2 % Iris-setosa
             5.5 3.5 1.3 0.2 4.9 3.1 1.5 0.1 % Iris-setosa
             4.4 3.0 1.3 0.2 5.1 3.4 1.5 0.2 % Iris-setosa
             5.0 3.5 1.3 0.3 4.5 2.3 1.3 0.3 % Iris-setosa
             4.4 3.2 1.3 0.2 5.0 3.5 1.6 0.6 % Iris-setosa
             5.1 3.8 1.9 0.4 4.8 3.0 1.4 0.3 % Iris-setosa
             5.1 3.8 1.6 0.2 4.6 3.2 1.4 0.2 % Iris-setosa
             5.3 3.7 1.5 0.2 5.0 3.3 1.4 0.2 % Iris-setosa
             7.0 3.2 4.7 1.4 6.4 3.2 4.5 1.5 % Iris-versicolor
             6.9 3.1 4.9 1.5 5.5 2.3 4.0 1.3 % Iris-versicolor
             6.5 2.8 4.6 1.5 5.7 2.8 4.5 1.3 % Iris-versicolor
             6.3 3.3 4.7 1.6 4.9 2.4 3.3 1.0 % Iris-versicolor
             6.6 2.9 4.6 1.3 5.2 2.7 3.9 1.4 % Iris-versicolor
             5.0 2.0 3.5 1.0 5.9 3.0 4.2 1.5 % Iris-versicolor
             6.0 2.2 4.0 1.0 6.1 2.9 4.7 1.4 % Iris-versicolor
             5.6 2.9 3.6 1.3 6.7 3.1 4.4 1.4 % Iris-versicolor
             5.6 3.0 4.5 1.5 5.8 2.7 4.1 1.0 % Iris-versicolor
             6.2 2.2 4.5 1.5 5.6 2.5 3.9 1.1 % Iris-versicolor
             5.9 3.2 4.8 1.8 6.1 2.8 4.0 1.3 % Iris-versicolor
             6.3 2.5 4.9 1.5 6.1 2.8 4.7 1.2 % Iris-versicolor]
```

```
6.4 2.9 4.3 1.3 6.6 3.0 4.4 1.4 % Iris-versicolor
6.8 2.8 4.8 1.4 6.7 3.0 5.0 1.7 % Iris-versicolor
6.0 2.9 4.5 1.5 5.7 2.6 3.5 1.0 % Iris-versicolor
5.5 2.4 3.8 1.1 5.5 2.4 3.7 1.0 % Iris-versicolor
5.8 2.7 3.9 1.2 6.0 2.7 5.1 1.6 % Iris-versicolor
5.4 3.0 4.5 1.5 6.0 3.4 4.5 1.6 % Iris-versicolor
6.7 3.1 4.7 1.5 6.3 2.3 4.4 1.3 % Iris-versicolor
5.6 3.0 4.1 1.3 5.5 2.5 4.0 1.3 % Iris-versicolor
5.5 2.6 4.4 1.2 6.1 3.0 4.6 1.4 % Iris-versicolor
5.8 2.6 4.0 1.2 5.0 2.3 3.3 1.0 % Iris-versicolor
5.6 2.7 4.2 1.3 5.7 3.0 4.2 1.2 % Iris-versicolor
5.7 2.9 4.2 1.3 6.2 2.9 4.3 1.3 % Iris-versicolor
5.1 2.5 3.0 1.1 5.7 2.8 4.1 1.3 % Iris-versicolor
6.3 3.3 6.0 2.5 5.8 2.7 5.1 1.9 % Iris-verginica
7.1 3.0 5.9 2.1 6.3 2.9 5.6 1.8 % Iris-verginica
6.5 3.0 5.8 2.2 7.6 3.0 6.6 2.1 % Iris-verginica
4.9 2.5 4.5 1.7 7.3 2.9 6.3 1.8 % Iris-verginica
6.7 2.5 5.8 1.8 7.2 3.6 6.1 2.5 % Iris-verginica
6.5 3.2 5.1 2.0 6.4 2.7 5.3 1.9 % Iris-verginica
6.8 3.0 5.5 2.1 5.7 2.5 5.0 2.0 % Iris-verginica
5.8 2.8 5.1 2.4 6.4 3.2 5.3 2.3 % Iris-verginica
6.5 3.0 5.5 1.8 7.7 3.8 6.7 2.2 % Iris-verginica
7.7 2.6 6.9 2.3 6.0 2.2 5.0 1.5 % Iris-verginica
6.9 3.2 5.7 2.3 5.6 2.8 4.9 2.0 % Iris-verginica
7.7 2.8 6.7 2.0 6.3 2.7 4.9 1.8 % Iris-verginica
6.7 3.3 5.7 2.1 7.2 3.2 6.0 1.8 % Iris-verginica
6.2 2.8 4.8 1.8 6.1 3.0 4.9 1.8 % Iris-verginica
6.4 2.8 5.6 2.1 7.2 3.0 5.8 1.6 % Iris-verginica
7.4 2.8 6.1 1.9 7.9 3.8 6.4 2.0 % Iris-verginica
6.4 2.8 5.6 2.2 6.3 2.8 5.1 1.5 % Iris-verginica
6.1 2.6 5.6 1.4 7.7 3.0 6.1 2.3 % Iris-verginica
6.3 3.4 5.6 2.4 6.4 3.1 5.5 1.8 % Iris-verginica
6.0 3.0 4.8 1.8 6.9 3.1 5.4 2.1 % Iris-verginica
6.7 3.1 5.6 2.4 6.9 3.1 5.1 2.3 % Iris-verginica
5.8 2.7 5.1 1.9 6.8 3.2 5.9 2.3 % Iris-verginica
6.7 3.3 5.7 2.5 6.7 3.0 5.2 2.3 % Iris-verginica
6.3 2.5 5.0 1.9 6.5 3.0 5.2 2.0 % Iris-verginica
6.2 3.4 5.4 2.3 5.9 3.0 5.1 1.8]; % Iris-verginica
```

```
% =====
% Filename: Iris_bp.m
% =====

echo off

disp(' =====')
disp(' Iris plant classification: back-propagation algorithm')
disp(' =====')

disp(' =====')
disp(' Reference: Negnevitsky, M., "Artificial Intelligence: A Guide to Intelligent')
disp(' Systems", 2nd edn. Addison Wesley, Harlow, England, 2005. ')
disp(' Sec. 9.4 Will a neural network work for my problem? ')
disp(' =====')

disp('✓')
disp(' Problem: The Iris plant data set contains 3 classes, and each class is represented ✓')
disp(' by 50 plants. A plant is characterised by its sepal length, sepal width, ✓')
disp(' petal length and petal width. A three-layer back-propagation network is ✓')
disp(' required to classify Iris plants. ✓')
disp('✓')
disp(' =====')

[iris_data] = Iris_data;
iris_data = (iris_data(:,1:4))';

% Massaged values for the Iris plant data set

for n=1:4;
    iris_inputs(n,:)=(iris_data(n,:)-min(iris_data(n,:)))/ ...
    (max(iris_data(n,:))-min(iris_data(n,:)));
end

iris_target1 = [1 0 0]'; setosa=find(iris_target1);
iris_target2 = [0 1 0]'; versicolor=find(iris_target2);
iris_target3 = [0 0 1]'; verginica=find(iris_target3);

for n=1:(50-1)
    iris_target1=[iris_target1 iris_target1(:,1)];
    iris_target2=[iris_target2 iris_target2(:,1)];
    iris_target3=[iris_target3 iris_target3(:,1)];
end

iris_targets = [iris_target1 iris_target2 iris_target3];
```

```
disp('Hit any key to randomly select input vectors to be used in training. ')
disp(' ')
pause

p=[]; t=[]; test_p=[]; test_t=[];

for n=1:150
    if rand(1)>1/3
        p=[p iris_inputs(:,n)];
        t=[t iris_targets(:,n)];
    else
        test_p=[test_p iris_inputs(:,n)];
        test_t=[test_t iris_targets(:,n)];
    end
end

[m n]=size(test_p);

disp(' ')
fprintf(1,' The training data set contains %0f elements.\n', (150-n));
fprintf(1,' The test data set contains %0f elements.\n', n);
disp(' ')

echo on

% Hit any key to define the network architecture.
pause

s1=5; % Five neurons in the hidden layer
s2=3; % Three neuron in the output layer

% Hit any key to create the network, initialise its weights and biases,
% and set up training parameters.
pause

rand('seed',1243);

net = newff([4.3 7.9; 2.0 4.4; 1.0 6.9; 0.1 2.5],[s1 s2],{'logsig'✓
'purelin'},'traingdx');

net.trainParam.show=20; % Number of epochs between showing the progress
net.trainParam.epochs=1000; % Maximum number of epochs
net.trainParam.goal=0.001; % Performance goal
net.trainParam.lr=0.1; % Learning rate
net.trainParam.lr_inc=1.05; % Learning rate increase multiplier
net.trainParam.lr_dec=0.7; % Learning rate decrease multiplier
net.trainParam.mc=0.9; % Momentum constant

% Hit any key to train the back-propagation network.
pause
```

```

net=train(net,p,t);

echo off

disp(' ')
fprintf(1,' Iris-setosa is represented by output:   %.0f \n',setosa);
fprintf(1,' Iris-versicolor is represented by output: %.0f \n',versicolor);
fprintf(1,' Iris-verginica is represented by output:   %.0f \n',verginica);

disp(' ')
disp(' Hit any key to test the network using the test data set.')
disp(' ')
pause

n_setosa=0; n_versicolor=0; n_verginica=0;
error_setosa=0; error_versicolor=0; error_verginica=0; error=0;

fprintf(' Sepal length  Sepal width  Petal length  Petal width  Desired output  Actual ✓
output  Error\n');

for i=1:n
    fprintf('          %.1f          %.1f          %.1f          %.1f',test_p(1,i),test_p(2,
i),test_p(3,i),test_p(4,i));
    a=compet(sim(net,test_p(:,i))); a=find(a);
    b=compet(test_t(:,i)); b=find(b);
    if b==1
        n_setosa=n_setosa+1;
        fprintf('          Iris-setosa          ');
        if abs(a-b)>0
            error_setosa=error_setosa+1;
            fprintf('%.0f          Yes\n',a);
        else
            fprintf('%.0f          No\n',a);
        end
    elseif b==2
        n_versicolor=n_versicolor+1;
        fprintf('          Iris-versicolor          ');
        if abs(a-b)>0
            error_versicolor=error_versicolor+1;
            fprintf('%.0f          Yes\n',a);
        else
            fprintf('%.0f          No\n',a);
        end
    else
        n_verginica=n_verginica+1;
        fprintf('          Iris-verginica          ');
        if abs(a-b)>0
            error_verginica=error_verginica+1;
            fprintf('%.0f          Yes\n',a);
        else
            fprintf('%.0f          No\n',a);
        end
    end
end

```

```

end
end

error=(error_setosa+error_versicolor+error_verginica)/n*100;

error_setosa=error_setosa/n_setosa*100;
error_versicolor=error_versicolor/n_versicolor*100;
error_verginica=error_verginica/n_verginica*100;

fprintf(1,' \n')
fprintf(1,' Iris-setosa recognition error:   %.2f \n',error_setosa);
fprintf(1,' Iris-versicolor recognition error: %.2f \n',error_versicolor);
fprintf(1,' Iris-verginica recognition error:   %.2f \n',error_verginica);
fprintf(1,' \n')
fprintf(1,' Total Iris plant recognition error: %.2f \n',error);
fprintf(1,' \n')

disp('end of Iris_bp.m')

```

```

%Edris Evolutionary NN
%inputs = 4
%weights = 16

%hidden = 4
%bias = 4
%weights = 12

%out = 3
%bias = 3

%total weights = 35
% function [msE] = myNNS(W)

% W=rand(100, 27);

echo off;
rng('default');
W=rand(100, 35);
W1 = W;

%input training data
[iris_data] = Iris_data;
iris_data = (iris_data(:,1:4)))';

for n=1:4; %scaling
    iris_inputs(n,:)=(iris_data(n,:)-min(iris_data(n,:)))/ ...
        (max(iris_data(n,:))-min(iris_data(n,:))));
end

%target training data
iris_target1 = [1 0 0]'; setosa=find(iris_target1);
iris_target2 = [0 1 0]'; versicolor=find(iris_target2);
iris_target3 = [0 0 1]'; verginica=find(iris_target3);

for n=1:(50-1)
    iris_target1=[iris_target1 iris_target1(:,1)];
    iris_target2=[iris_target2 iris_target2(:,1)];
    iris_target3=[iris_target3 iris_target3(:,1)];
end

iris_targets = [iris_target1 iris_target2 iris_target3];

p=[]; t=[]; test_p=[]; test_t=[]; setosaind=[]; versicolorind=[]; verginicaind=[]; amsE= ✓
[];

testc = 0;
for n=1:150
    if rand(1)>1/3 %training data
        p=[p iris_inputs(:,n)];
        t=[t iris_targets(:,n)];
    end
end

```

```

else %testing data
    testc = testc+1; %increment test index
    test_p=[test_p iris_inputs(:,n)];
    test_t=[test_t iris_targets(:,n)];
    if n<51
        setosaind=[setosaind n];
        endsetosa = testc; %index of test where setosaind ends
    elseif n<101
        versicolorind=[versicolorind n];
        endvcolor = testc; %index of test where versicolorind ends
    else
        verginicaind=[verginicaind n];
        endverg = testc; %index of test where verginicaind ends
    end
end
end

[a nsetosa] = size(setosaind)
[a nvcolor] = size(versicolorind)
[a nverg] = size(verginicaind)
% %P is input data for NN
% P = iris_data;
%
% %T is training output for NN
% T = vertcat(iris_target1', iris_target2', iris_target3')';

[m n]=size(test_p);

trainingsize = 150 - n;
testingsize = n;

disp(' ')
fprintf(1,' The training data set contains %.0f elements.\n', (150-n));
fprintf(1,' The test data set contains %.0f elements.\n',n);
disp(' ')

chromo = 1;
train = 1;
tic;
for train = 1:trainingsize;
    for chromo = 1:100;
        %weights (to, from)
        % [row 1 = from first layer
        % column 1 = to first layer]
        % to hidden[1,4] from input[1,3]
        % [
            W(1)                                W(2)                                W(3) ✓
W(4);          W(29) %hidden1                  W2'1                                W3'1 ✓
            % bias
W4'1;          h1=p(1,train)*W(chromo,1)+ p(2,train)*W(chromo,2)+ p(2,train)*W(chromo,3)+ p(4, ✓
train)*W(chromo,4)+W(chromo,29);
        z1= 1/(1+exp(-h1));
    end
end

```

```

W(8); % W(5) W(6) W(7) ✓
% W(30) %hidden2 W1'2 W2'2 W3'2 ✓
W4'2; bias
h2=p(1,train)*W(chromo,5)+ p(2,train)*W(chromo,6)+ p(2,train)*W(chromo,7)+ p(4, ✓
train)*W(chromo,8)+W(chromo,30);
z2= 1/(1+exp(-h2));

% W(9) W(10) W(11) W(12); W(31) %hidden3
% W1'3 W2'3 W3'3 W4'3; bias
h3=p(1,train)*W(chromo,9)+ p(2,train)*W(chromo,10)+ p(2,train)*W(chromo,11)+ p(4, ✓
train)*W(chromo,12)+W(chromo,31);
z3= 1/(1+exp(-h3));

% W(13) W(14) W(15) W(16); W(32) %hidden4
% W1'4 W2'4 W3'4 W4'4; bias
h4=p(1,train)*W(chromo,13)+ p(2,train)*W(chromo,14)+ p(2,train)*W(chromo,15)+ p( ✓
(4,train)*W(chromo,16)+W(chromo,32);
z4= 1/(1+exp(-h4));

%to output (layer 3) from hidden (layer 2)
% [W15 W25 W35 W45 ✓
T5; out1 % W(17) W(18) W(19) W(20) ✓
W(33); out1 = z1*W(chromo, 17) + z2*W(chromo, 18) + z3*W(chromo, 19) + z4*W(chromo, 20) + W ✓
(chromo, 33);

% W16 W26 W36 W46 ✓
T6; % W(21) W(22) W(23) W(24) ✓
W(34); out2 = z1*W(chromo, 21) + z2*W(chromo, 22) + z3*W(chromo, 23) + z4*W(chromo, 24) + W ✓
(chromo, 34);

% W17 W27 W37 W47 ✓
T7]; % W(25) W(26) W(27) W(28) ✓
W(35)]; out3 = z1*W(chromo, 25) + z2*W(chromo, 26) + z3*W(chromo, 27) + z4*W(chromo, 28) + W ✓
(chromo, 35);

%get simulated outputs Y from defined NN net with defined weights W,
%and input P
Y = [out1; out2; out3];

%Error = expected - simulated = T - Y
E(chromo,:) = t(:,train) - Y;
mse(chromo,:) = mse(E(chromo,:)); %mean squared error
end %end population

```

```

Wnext = zeros(100, 35); %next population
Esearch = msE; %used for selection to find smallest error

amsE = [amsE, train];
if train > 1
    if amsE(train) < 0.07
        break %stop training
    end
end

%selection order of E index (least E) chromosomes-----
%this array order least mse to worst mse indices
%which coorespond to W indecies
for i = 1:100;
    [val, ind] = min(Esearch);
    sel(i,1) = ind;
    Esearch(ind, 1) = 100;
end

% sel; %array of indecies of E order min - max

%crossover-----
pcross = 0.2;
Xind = randperm(40); %select order of top 40 indcies to be crossed
Xgene = rand(7,1); %probability or crossover per gene one gene per hidden per output
for n = 1:20;
    for gene = 1:7;
        if gene == 1; %hidden 1 1:4; 29
            if Xgene(gene, 1) < pcross; %cross
                Wnext( ( Xind(2*n-1)), 1:4) = W( sel(Xind(2*n)), 1:4);
                Wnext( ( Xind(2*n-1)), 29) = W( sel(Xind(2*n)), 29);
                Wnext( ( Xind(2*n)), 1:4) = W( sel(Xind(2*n-1)), 1:4);
                Wnext( ( Xind(2*n)), 29) = W( sel(Xind(2*n-1)), 29);
            else %copy
                Wnext( ( Xind(2*n-1)), 1:4) = W( sel(Xind(2*n-1)), 1:4);
                Wnext( ( Xind(2*n-1)), 29) = W( sel(Xind(2*n-1)), 29);
                Wnext( ( Xind(2*n)), 1:4) = W( sel(Xind(2*n)), 1:4);
                Wnext( ( Xind(2*n)), 29) = W( sel(Xind(2*n)), 29);
            end
        elseif gene == 2; %hidden 2 5:8; 30
            if Xgene(gene, 1) < pcross; %cross
                Wnext( ( Xind(2*n-1)), 5:8) = W( sel(Xind(2*n)), 5:8);
                Wnext( ( Xind(2*n-1)), 30) = W( sel(Xind(2*n)), 30);
                Wnext( ( Xind(2*n)), 5:8) = W( sel(Xind(2*n-1)), 5:8);
                Wnext( ( Xind(2*n)), 30) = W( sel(Xind(2*n-1)), 30);
            else %copy
                Wnext( ( Xind(2*n-1)), 5:8) = W( sel(Xind(2*n-1)), 5:8);
                Wnext( ( Xind(2*n-1)), 30) = W( sel(Xind(2*n-1)), 30);
                Wnext( ( Xind(2*n)), 5:8) = W( sel(Xind(2*n)), 5:8);
            end
        end
    end
end

```

```

Wnext( ( Xind(2*n)), 30) = W( sel(Xind(2*n)), 30);
end

elseif gene == 3; %hidden 3 9:12; 31;
if Xgene(gene, 1) < pcross; %cross
    Wnext( ( Xind(2*n-1)), 9:12) = W( sel(Xind(2*n)), 9:12);
    Wnext( ( Xind(2*n-1)), 31) = W( sel(Xind(2*n)), 31);
    Wnext( ( Xind(2*n)), 9:12) = W( sel(Xind(2*n-1)), 9:12);
    Wnext( ( Xind(2*n)), 31) = W( sel(Xind(2*n-1)), 31);
else %copy
    Wnext( ( Xind(2*n-1)), 9:12) = W( sel(Xind(2*n-1)), 9:12);
    Wnext( ( Xind(2*n-1)), 31) = W( sel(Xind(2*n-1)), 31);
    Wnext( ( Xind(2*n)), 9:12) = W( sel(Xind(2*n)), 9:12);
    Wnext( ( Xind(2*n)), 31) = W( sel(Xind(2*n)), 31);
end

elseif gene == 4; %hidden 4 13:16; 32
if Xgene(gene, 1) < pcross; %cross
    Wnext( ( Xind(2*n-1)), 13:16) = W( sel(Xind(2*n)), 13:16);
    Wnext( ( Xind(2*n-1)), 32) = W( sel(Xind(2*n)), 32);
    Wnext( ( Xind(2*n)), 13:16) = W( sel(Xind(2*n-1)), 13:16);
    Wnext( ( Xind(2*n)), 32) = W( sel(Xind(2*n-1)), 32);
else %copy
    Wnext( ( Xind(2*n-1)), 13:16) = W( sel(Xind(2*n-1)), 13:16);
    Wnext( ( Xind(2*n-1)), 32) = W( sel(Xind(2*n-1)), 32);
    Wnext( ( Xind(2*n)), 13:16) = W( sel(Xind(2*n)), 13:16);
    Wnext( ( Xind(2*n)), 32) = W( sel(Xind(2*n)), 32);
end

elseif gene == 5; %out 1 17:20; 33
if Xgene(gene, 1) < pcross; %cross
    Wnext( ( Xind(2*n-1)), 17:20) = W( sel(Xind(2*n)), 17:20);
    Wnext( ( Xind(2*n-1)), 33) = W( sel(Xind(2*n)), 33);
    Wnext( ( Xind(2*n)), 17:20) = W( sel(Xind(2*n-1)), 17:20);
    Wnext( ( Xind(2*n)), 33) = W( sel(Xind(2*n-1)), 33);
else %copy
    Wnext( ( Xind(2*n-1)), 17:20) = W( sel(Xind(2*n-1)), 17:20);
    Wnext( ( Xind(2*n-1)), 33) = W( sel(Xind(2*n-1)), 33);
    Wnext( ( Xind(2*n)), 17:20) = W( sel(Xind(2*n)), 17:20);
    Wnext( ( Xind(2*n)), 33) = W( sel(Xind(2*n)), 33);
end

elseif gene == 6; %out2 21:24; 34
if Xgene(gene, 1) < pcross; %cross
    Wnext( ( Xind(2*n-1)), 21:24) = W( sel(Xind(2*n)), 21:24);
    Wnext( ( Xind(2*n-1)), 34) = W( sel(Xind(2*n)), 34);
    Wnext( ( Xind(2*n)), 21:24) = W( sel(Xind(2*n-1)), 21:24);
    Wnext( ( Xind(2*n)), 34) = W( sel(Xind(2*n-1)), 34);
else %copy
    Wnext( ( Xind(2*n-1)), 21:24) = W( sel(Xind(2*n-1)), 21:24);
    Wnext( ( Xind(2*n-1)), 34) = W( sel(Xind(2*n-1)), 34);

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```

Wnext( ( Xind(2*n)), 21:24) = W( sel(Xind(2*n)), 21:24);
Wnext( ( Xind(2*n)), 34) = W( sel(Xind(2*n)), 34);
end

elseif gene == 7; %out3 25:28; 35
if Xgene(gene, 1) < pcross; %cross
    Wnext( ( Xind(2*n-1)), 25:28) = W( sel(Xind(2*n)), 25:28);
    Wnext( ( Xind(2*n-1)), 35) = W( sel(Xind(2*n)), 35);
    Wnext( ( Xind(2*n)), 25:28) = W( sel(Xind(2*n-1)), 25:28);
    Wnext( ( Xind(2*n)), 35) = W( sel(Xind(2*n-1)), 35);
else %copy
    Wnext( ( Xind(2*n-1)), 25:28) = W( sel(Xind(2*n-1)), 25:28);
    Wnext( ( Xind(2*n-1)), 35) = W( sel(Xind(2*n-1)), 35);
    Wnext( ( Xind(2*n)), 25:28) = W( sel(Xind(2*n)), 25:28);
    Wnext( ( Xind(2*n)), 35) = W( sel(Xind(2*n)), 35);
end

end

end

end

pmut = 0.001;
%mutation-----
for n = 41:100; %mutated top 60 genes
    if rand() < pmute; %MUTATE
        pos1 = randperm(35);
        pos = pos1(1);
        if pos < 35 %mutate some bits
            Wnext(n, 1:pos) = randn(1, 1:pos)+W(sel(n-40), 1:pos);
            Wnext(n, (pos+1):35);
        else %mutate all bits
            Wnext(n, 1:35) = randn(1, 1:35)+W(sel(n-40), 1:35);
        end
    end
end

%update next weights
W = Wnext;
end %end training
trainingt = toc;

Wbest = W(sel(1), :);
scorrect = 0; vccorrect = 0; vergcorrect = 0; nerror = 0;
serror = 0; vcerror = 0; vergerror = 0; nerror = 0;

disp(' ')
fprintf(1, ' Iris-setosa is represented by output: %0f \n', setosa);
fprintf(1, ' Iris-versicolor is represented by output: %0f \n', versicolor);
fprintf(1, ' Iris-verginica is represented by output: %0f \n', verginica);

```

```

disp(' ')
disp(' Hit any key to test the network using the test data set.')
disp(' ')
pause
% Testing-----
%use test_p & test_t
tic;
for tn = 1:testingsize
    %weights (to, from)
    % [row 1 = from first layer
    %   column 1 = to first layer]
    % to hidden[1,4] from input[1,3]
    % [ W(1) W(2) W(3)
W(4); W(29) %hidden1 W(2) W(3)
% W1'1 W2'1 W3'1
W4'1; bias
h1=test_p(1,tn)*Wbest(1,1)+ test_p(2,tn)*Wbest(1,2)+ test_p(2,tn)*Wbest(1,3)+ test_p
(4,tn)*Wbest(1,4)+Wbest(1,29);
z1= 1/(1+exp(-h1));

% W(8); W(30) W(5) W(6) W(7)
% W1'2 W2'2 W3'2
W4'2; bias
h2=test_p(1,tn)*Wbest(1,5)+ test_p(2,tn)*Wbest(1,6)+ test_p(2,tn)*Wbest(1,7)+ test_p
(4,tn)*Wbest(1,8)+Wbest(1,3);
z2= 1/(1+exp(-h2));

% W(12); W(9) W(10) W(11)
% W1'3 W2'3 W3'3 W4'3; bias
h3=test_p(1,tn)*Wbest(1,9)+ test_p(2,tn)*Wbest(1,10)+ test_p(2,tn)*Wbest(1,11)+
test_p(4,tn)*Wbest(1,12)+Wbest(1,31);
z3= 1/(1+exp(-h3));

% W(16); W(13) W(14) W(15)
% W1'4 W2'4 W3'4 W4'4;
h4=test_p(1,tn)*Wbest(1,13)+ test_p(2,tn)*Wbest(1,14)+ test_p(2,tn)*Wbest(1,15)+
test_p(4,tn)*Wbest(1,16)+Wbest(1,32);
z4= 1/(1+exp(-h4));

%to output (layer 3) from hidden (layer 2)
% W15 W25 W35 W45
T5; out1 W(17) W(18) W(19) W(20) W
(33); out1 = z1*Wbest(1, 17) + z2*Wbest(1, 18)+ z3*Wbest(1, 19)+ z4*Wbest(1, 20)+ Wbest(1,
33);

```

```

% W16 W26 W36 W46
T6; W(21) W(22) W(23) W(24) W
(34); out2 = z1*Wbest(1, 21) + z2*Wbest(1, 22)+ z3*Wbest(1, 23)+ z4*Wbest(1, 24)+ Wbest(1,
34);

% W17 W27 W37 W47
T7]; W(25) W(26) W(27) W(28) W
(35)]; out3 = z1*Wbest(1, 25) + z2*Wbest(1, 26)+ z3*Wbest(1, 27)+ z4*Wbest(1, 28)+ Wbest(1,
35);

%get simulated outputs Y from defined NN net with defined weights W,
%and input P
if max([out1 out2 out3]) == out1;
    Yf = [1; 0; 0];
elseif max([out1, out2, out3]) == out2;
    Yf = [0; 1; 0];
else max([out1, out2, out3]) == out3;
    Yf = [0; 0; 1];
end
Yf

% if Yf(1) == test_t(1,tn) && Yf(2) == test_t(2,tn) && Yf(3) == test_t(3,tn) && tn <=
endsetosa
% scorrect = scorrect + 1;
% elseif Yf(1) == test_t(1,tn) && Yf(2) == test_t(2,tn) && Yf(3) == test_t(3,tn) &&
tn <= endvcolor
% vccorrect = vccorrect + 1;
% elseif Yf(1) == test_t(1,tn) && Yf(2) == test_t(2,tn) && Yf(3) == test_t(3,tn) &&
tn <= endverg
% vergcorrect = vergcorrect + 1;
% else nerror = nerror + 1;
% end
if tn > endvcolor && (Yf(1) ~= test_t(1,tn) || Yf(2) ~= test_t(2,tn) || Yf(3) ~=
test_t(3,tn))
    vergererror = vergererror + 1;
elseif tn > endsetosa && (Yf(1) ~= test_t(1,tn) || Yf(2) ~= test_t(2,tn) || Yf(3) ~=
test_t(3,tn))
    vcerror = vcerror + 1;
elseif tn <= endsetosa && (Yf(1) ~= test_t(1,tn) || Yf(2) ~= test_t(2,tn) || Yf(3) ~=
test_t(3,tn))
    serror = serror + 1;
% else nerror = nerror + 1;
% end
Y = [out1; out2; out3];

%Error = expected - simulated = T - Y
E(tn,:) = test_t(:,tn) - Y;

```

```

mse(E(tn,:)) = mse(E(tn,:)); %mean squared error
end
testingt = toc;

error = error/nsetosa*100;
vcerror = vcerror/nvcolor*100;
vergererror = vergererror/nverg*100;
nerror = nerror/testingsize*100;
terror = error + vergererror + vcerror + serror;

% error = abs(scorrect-nsetosa)/nsetosa*100;
% vcerror = abs(vccorrect-nvcolor)/nvcolor*100;
% vergererror = abs(vergcorrect-nverg)/nverg*100;
% nerror = abs(nerror-testingsize)/testingsize*100;
% terror = error + vergererror + vcerror + serror;

fprintf(1,' Training time: %2f \n', trainingt);
fprintf(1,' Testing time: %2f \n', testingt);
fprintf(1,' \n')
fprintf(1,' Iris-setosa recognition error: %2f \n',serror);
fprintf(1,' Iris-versicolor recognition error: %2f \n',vcerror);
fprintf(1,' Iris-verginica recognition error: %2f \n',vergererror);
fprintf(1,' missclassification error: %2f \n',nerror);
fprintf(1,' \n')
fprintf(1,' Total Iris plant recognition error: %2f \n',terror);
fprintf(1,' \n')

```

```

%Edris Evolutionary NN
%inputs = 4
%weights = 20

%hidden = 5
%bias = 5
%weights = 15

%out = 3
%bias = 3

%total weights = 43
% function [mse] = myNN5(W)

% W=rand(100, 27);
% W=rand(100, 35);
echo off;
rng('default');
W=randn(100, 43);
W1 = W;

%input training data
[iris_data] = iris_data;
iris_data = (iris_data(:,1:4))';

for n=1:4;%scaling
    iris_inputs(n,:)=(iris_data(n,:)-min(iris_data(n,:)))/ ...
        (max(iris_data(n,:))-min(iris_data(n,:))));
end

%target training data
iris_target1 = [1 0 0]'; setosa=find(iris_target1);
iris_target2 = [0 1 0]'; versicolor=find(iris_target2);
iris_target3 = [0 0 1]'; verginica=find(iris_target3);

for n=1:(50-1)
    iris_target1=[iris_target1 iris_target1(:,1)];
    iris_target2=[iris_target2 iris_target2(:,1)];
    iris_target3=[iris_target3 iris_target3(:,1)];
end

iris_targets = [iris_target1 iris_target2 iris_target3];

p=[]; t=[]; test_p=[]; test_t=[]; setosaind=[]; versicolorind=[]; verginicaind=[]; amsE=
[];

testc = 0;
for n=1:150
    if rand(1)>1/3 %trainging data
        p=[p iris_inputs(:,n)];
        t=[t iris_targets(:,n)];
    end
end

```

```

else %testing data
    testc = testc+1; %increment test index
    test_p=[test_p iris_inputs(:,n)];
    test_t=[test_t iris_targets(:,n)];
    if n<51
        setosaind=[setosaind n];
        endsetosa = testc; %index of test where setosaind ends
    elseif n<101
        versicolorind=[versicolorind n];
        endvcolor = testc; %index of test where versicolorind ends
    else
        verginicaind=[verginicaind n];
        endverg = testc; %index of test where verginicaind ends
    end
end
end

[a nsetosa] = size(setosaind)
[a nvcolor] = size(versicolorind)
[a nverg] = size(verginicaind)
% P is input data for NN
% P = iris_data;
%
% T is training output for NN
% T = vertcat(iris_target1', iris_target2', iris_target3');

[m n]=size(test_p);

trainingsize = 150 - n;
testingsize = n;

disp(' ')
fprintf(1,' The training data set contains %.0f elements.\n', (150-n));
fprintf(1,' The test data set contains %.0f elements.\n',n);
disp(' ')

chromo = 1;
train = 1;
tic;
for train = 1:trainingsize;
    for chromo = 1:100;
        %weights (to, from)
        % [row 1 = from first layer
        %   column 1 = to first layer]
        % to hidden[1,5] from input[1,3]
        % [W(1) W(2) W(3) W(4); W(36)bias %hidden1
        %   W(1' W(2' W(3' W(4';
        h1=p(1,train)*W(chromo,1)+ p(2,train)*W(chromo,2)+ p(2,train)*W(chromo,3)+ p(4,
train)*W(chromo,4)+W(chromo,36);
        z1 = 1/(1+exp(-h1));

        % W(5) W(6) W(7) W(8); %hidden2

```

```

% W1'2 W2'2 W3'2 W4'2;
h2=p(1,train)*W(chromo,5)+ p(2,train)*W(chromo,6)+ p(2,train)*W(chromo,7)+ p(4,
train)*W(chromo,8)+W(chromo,37);
z2= 1/(1+exp(-h2));

% W(9) W(10) W(11) W(12); %hidden3
% W1'3 W2'3 W3'3 W4'3;
h3=p(1,train)*W(chromo,9)+ p(2,train)*W(chromo,10)+ p(2,train)*W(chromo,11)+ p(4,
train)*W(chromo,12)+W(chromo,38);
z3= 1/(1+exp(-h3));

% W(13) W(14) W(15) W(16) %hidden4
% W1'4 W2'4 W3'4 W4'4;
h4=p(1,train)*W(chromo,13)+ p(2,train)*W(chromo,14)+ p(2,train)*W(chromo,15)+ p
(4,train)*W(chromo,16)+W(chromo,39);
z4= 1/(1+exp(-h4));

% W(17) W(18) W(19) W(20); %hidden5
% W1'5 W2'5 W3'5
h5=p(1,train)*W(chromo,17)+ p(2,train)*W(chromo,18)+ p(2,train)*W(chromo,19)+ p
(4,train)*W(chromo,20)+W(chromo,40);
z5= 1/(1+exp(-h5));

%to output (layer 3) from hidden (layer 2)
% W16 W26 W36 W46
W56 % T6; out1 W(22) W(23) W(24)
% W(21) W(41);
W(25) out1 = z1*W(chromo, 21) + z2*W(chromo, 22)+ z3*W(chromo, 23)+ z4*W(chromo, 24)+
z5*W(chromo, 35)+ W(chromo, 41);

% W17 W27 W37 W47
W57 % T7; W(26) W(27) W(28) W(29)
% W(42); out2
W(30) out2 = z1*W(chromo, 26) + z2*W(chromo, 27)+ z3*W(chromo, 28)+ z4*W(chromo, 29)+
z5*W(chromo, 30)+ W(chromo, 42);

% W18 W28 W38 W48 W58 T8; W(32) W(33) W(34)
W(35) % W(31) W(43);
out3 = z1*W(chromo, 31) + z2*W(chromo, 32)+ z3*W(chromo, 33)+ z4*W(chromo, 34)+
z5*W(chromo, 35)+ W(chromo, 43);

%get simulated outputs Y from defined NN net with defined weights W,
%and input P
Y = [out1; out2; out3];

>Error = expected - simulated = T - Y
E(chromo,:) = t(:,train) - Y;

```

```

mse(E(chromo,:)) = mse(E(chromo,:)); %mean squared error

end %end population

amsE = [amsE, train];
if train > 1
    if amsE(train) < 0.07
        break %stop training
    end
end

Wnext = zeros(100, 43); %next population
Esearch = mseE; %used for selection to find smallest error

%selection order of E index (least E) chromosomes-----
%this array order least mse to worst mse indices
%which coorespond to W indices
for i = 1:100;
    [val, ind] = min(Esearch);
    sel(i,1) = ind;
    Esearch(ind, 1) = 100;
end
% sel; %array of indices of E order min - max

%-----crossover-----
pcross = 0.2;
Wnext = randperm(40); %select order of top 40 indices to be crossed
Xgene = rand(8,14); %probability or crossover per gene
for n = 1:20;
    for gene = 1:8;
        if gene == 1; %hidden 1 1:4; 36
            if Xgene(gene, 1) < pcross; %cross
                Wnext( ( Xind(2*n-1)), 1:4) = W( sel(Xind(2*n)), 1:4);
                Wnext( ( Xind(2*n-1)), 36) = W( sel(Xind(2*n)), 36);
                Wnext( ( Xind(2*n)), 1:4) = W( sel(Xind(2*n-1)), 1:4);
                Wnext( ( Xind(2*n)), 36) = W( sel(Xind(2*n-1)), 36);
            else %copy
                Wnext( ( Xind(2*n-1)), 1:4) = W( sel(Xind(2*n-1)), 1:4);
                Wnext( ( Xind(2*n-1)), 36) = W( sel(Xind(2*n-1)), 36);
                Wnext( ( Xind(2*n)), 1:4) = W( sel(Xind(2*n)), 1:4);
                Wnext( ( Xind(2*n)), 36) = W( sel(Xind(2*n)), 36);
            end
        elseif gene == 2; %hidden 2 5:8; 37
            if Xgene(gene, 1) < pcross; %cross
                Wnext( ( Xind(2*n-1)), 5:8) = W( sel(Xind(2*n)), 5:8);
                Wnext( ( Xind(2*n-1)), 37) = W( sel(Xind(2*n)), 37);
                Wnext( ( Xind(2*n)), 5:8) = W( sel(Xind(2*n-1)), 5:8);
                Wnext( ( Xind(2*n)), 37) = W( sel(Xind(2*n-1)), 37);
            else %copy
                Wnext( ( Xind(2*n-1)), 5:8) = W( sel(Xind(2*n-1)), 5:8);

```

```

Wnext( Xind(2*n-1), 37) = W( sel(Xind(2*n-1)), 37);
Wnext( Xind(2*n), 5:8) = W( sel(Xind(2*n)), 5:8);
Wnext( Xind(2*n), 37) = W( sel(Xind(2*n)), 37);

end

elseif gene == 3;    %hidden 3 9:12; 38;
if Xgene(gene, 1) < pcross; %cross
    Wnext( Xind(2*n-1), 9:12) = W( sel(Xind(2*n)), 9:12);
    Wnext( Xind(2*n-1), 38) = W( sel(Xind(2*n)), 38);
    Wnext( Xind(2*n), 9:12) = W( sel(Xind(2*n-1)), 9:12);
    Wnext( Xind(2*n), 38) = W( sel(Xind(2*n-1)), 38);
else %copy
    Wnext( Xind(2*n-1), 9:12) = W( sel(Xind(2*n-1)), 9:12);
    Wnext( Xind(2*n-1), 38) = W( sel(Xind(2*n-1)), 38);
    Wnext( Xind(2*n), 9:12) = W( sel(Xind(2*n)), 9:12);
    Wnext( Xind(2*n), 38) = W( sel(Xind(2*n)), 38);
end

elseif gene == 4;    %hidden 4 13:16; 39
if Xgene(gene, 1) < pcross; %cross
    Wnext( Xind(2*n-1), 13:16) = W( sel(Xind(2*n)), 13:16);
    Wnext( Xind(2*n-1), 39) = W( sel(Xind(2*n)), 39);
    Wnext( Xind(2*n), 13:16) = W( sel(Xind(2*n-1)), 13:16);
    Wnext( Xind(2*n), 39) = W( sel(Xind(2*n-1)), 39);
else %copy
    Wnext( Xind(2*n-1), 13:16) = W( sel(Xind(2*n-1)), 13:16);
    Wnext( Xind(2*n-1), 39) = W( sel(Xind(2*n-1)), 39);
    Wnext( Xind(2*n), 13:16) = W( sel(Xind(2*n)), 13:16);
    Wnext( Xind(2*n), 39) = W( sel(Xind(2*n)), 39);
end

elseif gene == 5;    %hidden 5 17:20; 40
if Xgene(gene, 1) < pcross; %cross
    Wnext( Xind(2*n-1), 17:20) = W( sel(Xind(2*n)), 17:20);
    Wnext( Xind(2*n-1), 40) = W( sel(Xind(2*n)), 40);
    Wnext( Xind(2*n), 17:20) = W( sel(Xind(2*n-1)), 17:20);
    Wnext( Xind(2*n), 40) = W( sel(Xind(2*n-1)), 40);
else %copy
    Wnext( Xind(2*n-1), 17:20) = W( sel(Xind(2*n-1)), 17:20);
    Wnext( Xind(2*n-1), 40) = W( sel(Xind(2*n-1)), 40);
    Wnext( Xind(2*n), 17:20) = W( sel(Xind(2*n)), 17:20);
    Wnext( Xind(2*n), 40) = W( sel(Xind(2*n)), 40);
end

elseif gene == 6;    %out1 21:25; 41
if Xgene(gene, 1) < pcross; %cross
    Wnext( Xind(2*n-1), 21:25) = W( sel(Xind(2*n)), 21:25);
    Wnext( Xind(2*n-1), 41) = W( sel(Xind(2*n)), 41);
    Wnext( Xind(2*n), 21:25) = W( sel(Xind(2*n-1)), 21:25);
    Wnext( Xind(2*n), 41) = W( sel(Xind(2*n-1)), 41);
else %copy

```



```

Wnext( Xind(2*n-1)), 21:25) = W( sel(Xind(2*n-1)), 21:25);
Wnext( Xind(2*n-1)), 41) = W( sel(Xind(2*n-1)), 41);
Wnext( Xind(2*n)), 21:25) = W( sel(Xind(2*n)), 21:25);
Wnext( Xind(2*n)), 41) = W( sel(Xind(2*n)), 41);
end

elseif gene == 7; %out2 26:30; 42
if Xgene(gene, 1) < pcross; %cross
Wnext( Xind(2*n-1)), 26:30) = W( sel(Xind(2*n)), 26:30);
Wnext( Xind(2*n-1)), 42) = W( sel(Xind(2*n)), 42);
Wnext( Xind(2*n)), 26:30) = W( sel(Xind(2*n-1)), 26:30);
Wnext( Xind(2*n)), 42) = W( sel(Xind(2*n-1)), 42);
else %copy
Wnext( Xind(2*n-1)), 26:30) = W( sel(Xind(2*n-1)), 26:30);
Wnext( Xind(2*n-1)), 42) = W( sel(Xind(2*n-1)), 42);
Wnext( Xind(2*n)), 26:30) = W( sel(Xind(2*n)), 26:30);
Wnext( Xind(2*n)), 42) = W( sel(Xind(2*n)), 42);
end

else % gene == 8 out3 31:35; 43
if Xgene(gene, 1) < pcross; %cross
Wnext( Xind(2*n-1)), 31:35) = W( sel(Xind(2*n)), 31:35);
Wnext( Xind(2*n-1)), 43) = W( sel(Xind(2*n)), 43);
Wnext( Xind(2*n)), 31:35) = W( sel(Xind(2*n-1)), 31:35);
Wnext( Xind(2*n)), 43) = W( sel(Xind(2*n-1)), 43);
else %copy
Wnext( Xind(2*n-1)), 31:35) = W( sel(Xind(2*n-1)), 31:35);
Wnext( Xind(2*n-1)), 43) = W( sel(Xind(2*n-1)), 43);
Wnext( Xind(2*n)), 31:35) = W( sel(Xind(2*n)), 31:35);
Wnext( Xind(2*n)), 43) = W( sel(Xind(2*n)), 43);
end
end
end

pmut = 0.001;
%mutation-----
for n = 41:100; %mutated top 60 genes
if rand() < pmut; %MUTATE
pos1 = randperm(43);
pos = pos1(1);
if pos < 43 %mutate some bits
Wnext(n, 1:pos) = randn(1, 1:pos)+W(sel(n-40), 1:pos);
Wnext(n, (pos+1):43);
else %mutate all bits
Wnext(n, 1:43) = randn(1, 1:43)+W(sel(n-40), 1:43);
end
end
else
Wnext(n, 1:43) = W(sel(n-40), 1:43);
end
end
end

```

```

%update next weights
W = Wnext;
end %end training
trainingt = toc;

Wbest = W(sel(1), :);
scorrect = 0; vccorrect = 0; vergcorrect = 0; nerror = 0;
serror = 0; vcerror = 0; vergererror = 0; nerror = 0;

disp(' ')
fprintf(1, ' Iris-setosa is represented by output: %0f \n', setosa);
fprintf(1, ' Iris-versicolor is represented by output: %0f \n', versicolor);
fprintf(1, ' Iris-verginica is represented by output: %0f \n', verginica);

disp(' ')
disp(' Hit any key to test the network using the test data set.')
disp(' ')
pause
% Testing-----
%use test_p & test_t
tic;
for tn = 1:testingsize
%weights (to, from)
% [row 1 = from first layer
% column 1 = to first layer]
% to hidden[1,4] from input[1,3]
% W(1) W(2) W(3) W(4); W(36)bias %hidden1
% W1'1 W2'1 W3'1 W4'1;
h1=test_p(1,tn)*Wbest(1,1)+ test_p(2,tn)*Wbest(1,2)+ test_p(2,tn)*Wbest(1,3)+ test_p
(4,tn)*Wbest(1,4)+Wbest(1,36);
z1 = 1/(1+exp(-h1));

% W(5) W(6) W(7) W(8); %hidden2
% W1'2 W2'2 W3'2 W4'2;
h2=test_p(1,tn)*Wbest(1,5)+ test_p(2,tn)*Wbest(1,6)+ test_p(2,tn)*Wbest(1,7)+ test_p
(4,tn)*Wbest(1,8)+Wbest(1,37);
z2 = 1/(1+exp(-h2));

% W(9) W(10) W(11) W(12); %hidden3
% W1'3 W2'3 W3'3 W4'3;
h3=test_p(1,tn)*Wbest(1,9)+ test_p(2,tn)*Wbest(1,10)+ test_p(2,tn)*Wbest(1,11)+
test_p(4,tn)*Wbest(1,12)+Wbest(1,38);
z3 = 1/(1+exp(-h3));

% W(13) W(14) W(15) W(16) %hidden4
% W1'4 W2'4 W3'4 W4'4;
h4=test_p(1,tn)*Wbest(1,13)+ test_p(2,tn)*Wbest(1,14)+ test_p(2,tn)*Wbest(1,15)+
test_p(4,tn)*Wbest(1,16)+Wbest(1,39);
z4 = 1/(1+exp(-h4));

```

```

% W(17) W(18) W(19) W(20)); %hidden5
% W1'5 W2'5 W3'5
W4'5]
h5=test_p(1,tn)*Wbest(1,17)+ test_p(2,tn)*Wbest(1,18)+ test_p(2,tn)*Wbest(1,19)+
test_p(4,tn)*Wbest(1,20)+Wbest(1,40);
z5 = 1/(1+exp(-h5));

%to output (layer 3) from hidden (layer 2)
% W16 W26 W36 W46 W56
T6; out1
% W(21) W(22) W(23) W(24) W
(25)
out1 = z1*Wbest(1, 21) + z2*Wbest(1, 22)+ z3*Wbest(1, 23)+ z4*Wbest(1, 24)+ z5*Wbest
(1, 35)+ Wbest(1, 41);

% W17 W27 W37 W47
W57 T7;
% W(26) W(27) W(28) W(29)
W(30)
out2 = z1*Wbest(1, 26) + z2*Wbest(1, 27)+ z3*Wbest(1, 28)+ z4*Wbest(1, 29)+ z5*Wbest
(1, 30)+ Wbest(1, 42);

% W18 W28 W38 W48 W58 T8];
% W(31) W(32) W(33) W(34)
W(35)
out3 = z1*Wbest(1, 31) + z2*Wbest(1, 32)+ z3*Wbest(1, 33)+ z4*Wbest(1, 34)+ z5*Wbest
(1, 35)+ Wbest(1, 43);

%get simulated outputs Y from defined NN net with defined weights W,
%and input P
if max([out1 out2 out3]) == out1;
Yf = [1; 0; 0];
elseif max([out1, out2, out3]) == out2;
Yf = [0; 1; 0];
else % max([out1, out2, out3]) == out3;
Yf = [0; 0; 1];
end
Yf

% if Yf(1) == test_t(1,tn) && Yf(2) == test_t(2,tn) && Yf(3) == test_t(3,tn) && tn <=
endsetosa
% scorrect = scorrect + 1;
% elseif Yf(1) == test_t(1,tn) && Yf(2) == test_t(2,tn) && Yf(3) == test_t(3,tn) &&
tn <= endvcolor
% vccorrect = vccorrect + 1;
% elseif Yf(1) == test_t(1,tn) && Yf(2) == test_t(2,tn) && Yf(3) == test_t(3,tn) &&
tn <= endverg
% vergcorrect = vergcorrect + 1;
% else nerror = nerror + 1;
% end
if tn > endvcolor && (Yf(1) ~= test_t(1,tn) || Yf(2) ~= test_t(2,tn) || Yf(3) ~=

```

```

test_t(3,tn))
vergererror = vergererror + 1;
elseif tn > endsetosa && (Yf(1) ~= test_t(1,tn) || Yf(2) ~= test_t(2,tn) || Yf(3) ~=
test_t(3,tn))
vcerror = vcerror + 1;
elseif tn <= endsetosa && (Yf(1) ~= test_t(1,tn) || Yf(2) ~= test_t(2,tn) || Yf(3) ~=
test_t(3,tn))
serror = serror + 1;
% else nerror = nerror + 1;
end
Y = [out1; out2; out3];

%Error = expected - simulated = T - Y
E(tn,:) = test_t(:,tn) - Y;
mseE(tn,:) = mse(E(tn,:)); %mean squared error

end
testingt = toc;

serror = serror/nsetosa*100;
vcerror = vcerror/nvcolor*100;
vergererror = vergererror/nverg*100;
nerror = nerror/testingsize*100;
terror = nerror + vergererror + vcerror + serror;

% serror = abs(scurrect-nsetosa)/nsetosa*100;
% vcerror = abs(vccorrect-nvcolor)/nvcolor*100;
% vergererror = abs(vergcorrect-nverg)/nverg*100;
% nerror = abs(nerror-testingsize)/testingsize*100;
% terror = nerror + vergererror + vcerror + serror;

fprintf(1, ' Training time: %2f \n', trainingt);
fprintf(1, ' Testing time: %2f \n', testingt);
fprintf(1, ' \n')
fprintf(1, ' Iris-setosa recognition error: %2f \n', serror);
fprintf(1, ' Iris-versicolor recognition error: %2f \n', vcerror);
fprintf(1, ' Iris-verginica recognition error: %2f \n', vergererror);
fprintf(1, ' missclassification error: %2f \n', nerror);
fprintf(1, ' \n')
fprintf(1, ' Total Iris plant recognition error: %2f \n', terror);
fprintf(1, ' \n')

```